

# Diffraction geometry parameterisation and refinement

David Waterman

CCP4

DIALS-3 @ LBNL

- See Graeme's slides for overview
- Input diffraction spot indices, their centroids and estimated uncertainties ( $h, k, l; X, Y, \phi; \sigma_x, \sigma_y, \sigma_\phi$ )
- Use all (useful) data available to refine a model to reduce rmsd of predicted centroids
- Global refinement helps to recover from poorly defined parameters in local  $\phi$  window
- More physically meaningful: avoids mopping up of effects by correlated parameters and therefore obtains realistic parameter values
- Refine profile parameters separately
- Potential second round with improved centroid observations

# Lessons from LURE

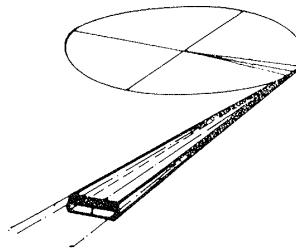
D.T. Thomas

Centre National de la Recherche Scientifique  
Université Paris-Sud

## Laboratoire pour l'Utilisation du Rayonnement Electromagnétique

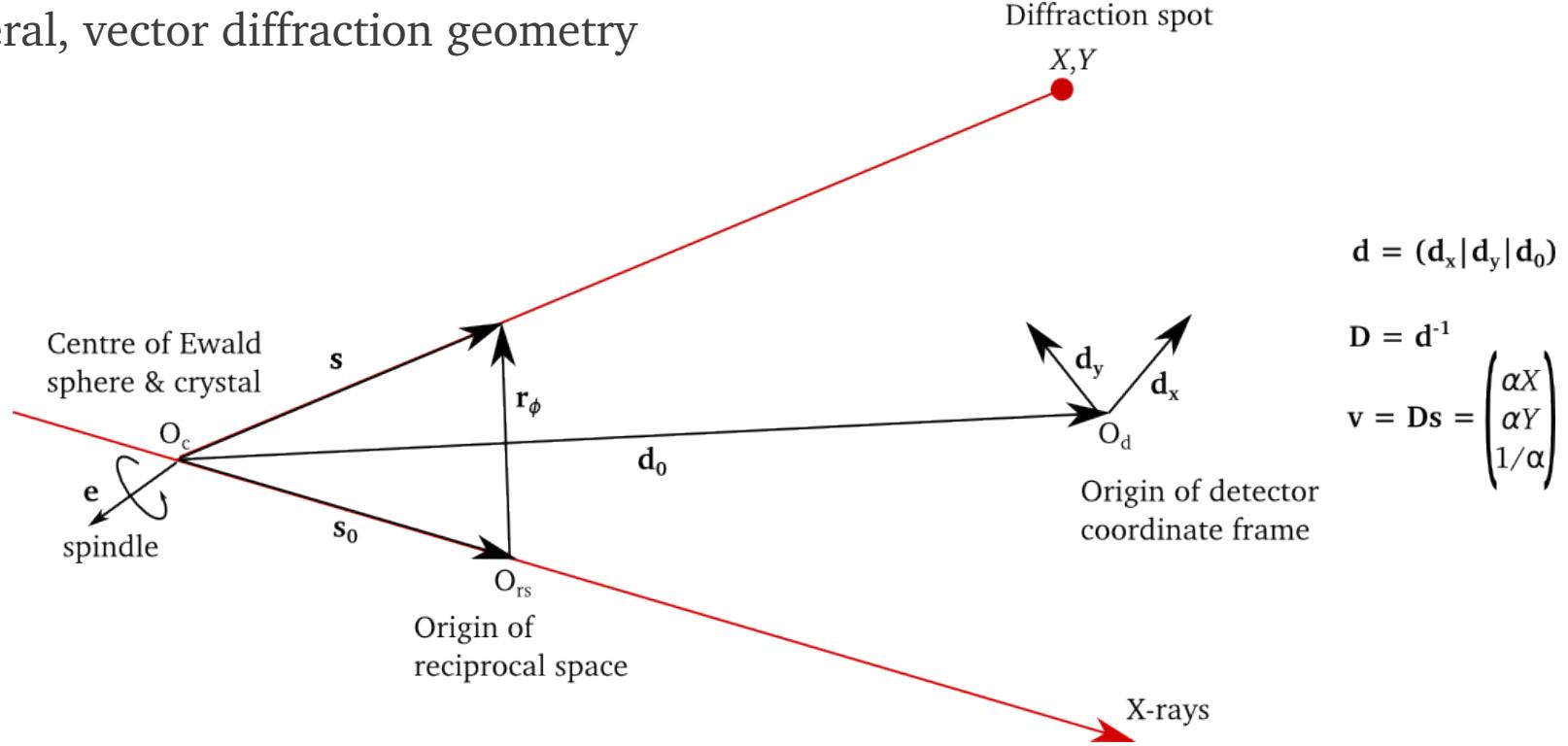
Proceedings  
of the EEC Cooperative Workshop  
on Position-Sensitive Detector Software  
(Phases I & II)  
held at L.U.R.E. from May 26 to June 7, 1986.

L. U. R. E.  
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- The description of diffraction geometry should be generalised and abstracted
- Do not depend on arbitrary reference frame conventions or hardware-specific features
- Parameterisation should be modular and extensible
- This is all 'obvious', but is easier to do from the ground up, with the toolbox approach

- General, vector diffraction geometry



- The detector abstract frame is a hardware-independent adapter
- Positional corrections can be accounted for in the mm-to-px mapping function

For refinement we want at least the first derivatives of predicted centroids

$$\frac{\partial \phi}{\partial p} = -\frac{\frac{\partial \mathbf{r}_\phi}{\partial p} \cdot \mathbf{s} + \mathbf{r}_\phi \cdot \frac{\partial \mathbf{s}_0}{\partial p}}{(\mathbf{e} \times \mathbf{r}_\phi) \cdot \mathbf{s}_0}$$

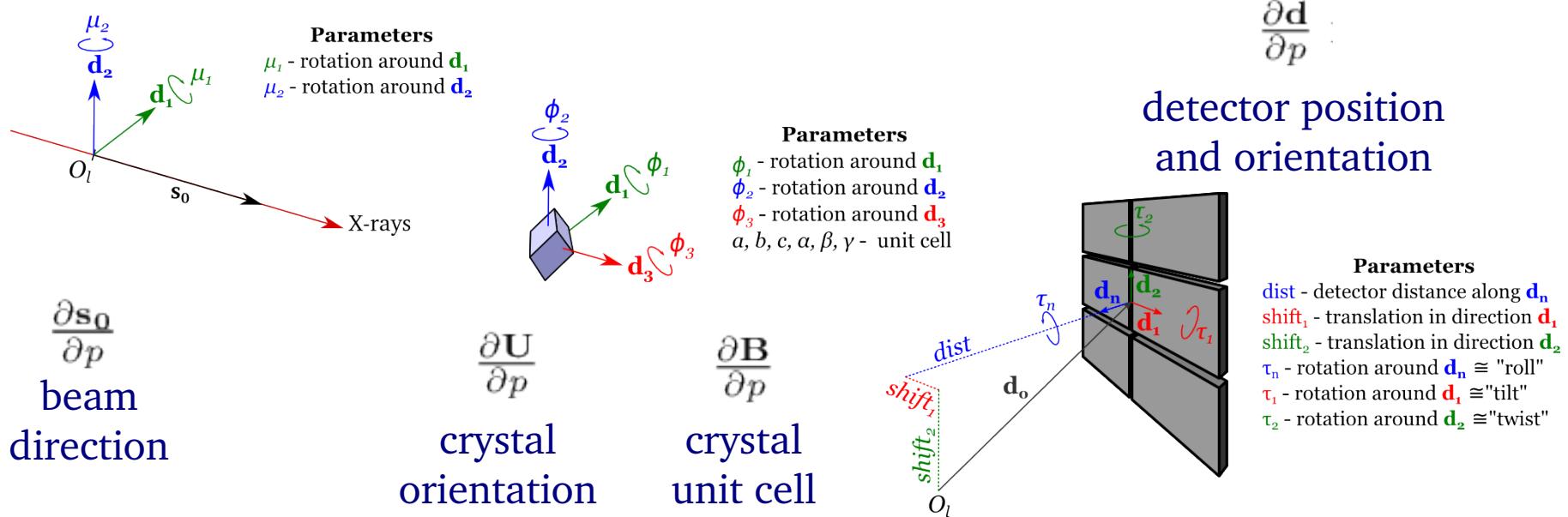
$$\frac{d\mathbf{v}}{dp} = -\mathbf{D} \frac{\partial \mathbf{d}}{\partial p} \mathbf{v} + \mathbf{D} \left[ \frac{\partial \mathbf{r}_\phi}{\partial p} + (\mathbf{e} \times \mathbf{r}_\phi) \frac{\partial \phi}{\partial p} + \frac{\partial \mathbf{s}_0}{\partial p} \right]$$

Neatly, these are factored into independent models

$$\begin{array}{ccc} \frac{\partial \mathbf{d}}{\partial p} & \frac{\partial \mathbf{r}_\phi}{\partial p} & \frac{\partial \mathbf{s}_0}{\partial p} \\ \text{detector} & \text{crystal} & \text{beam} \\ & & \text{direction} \end{array}$$

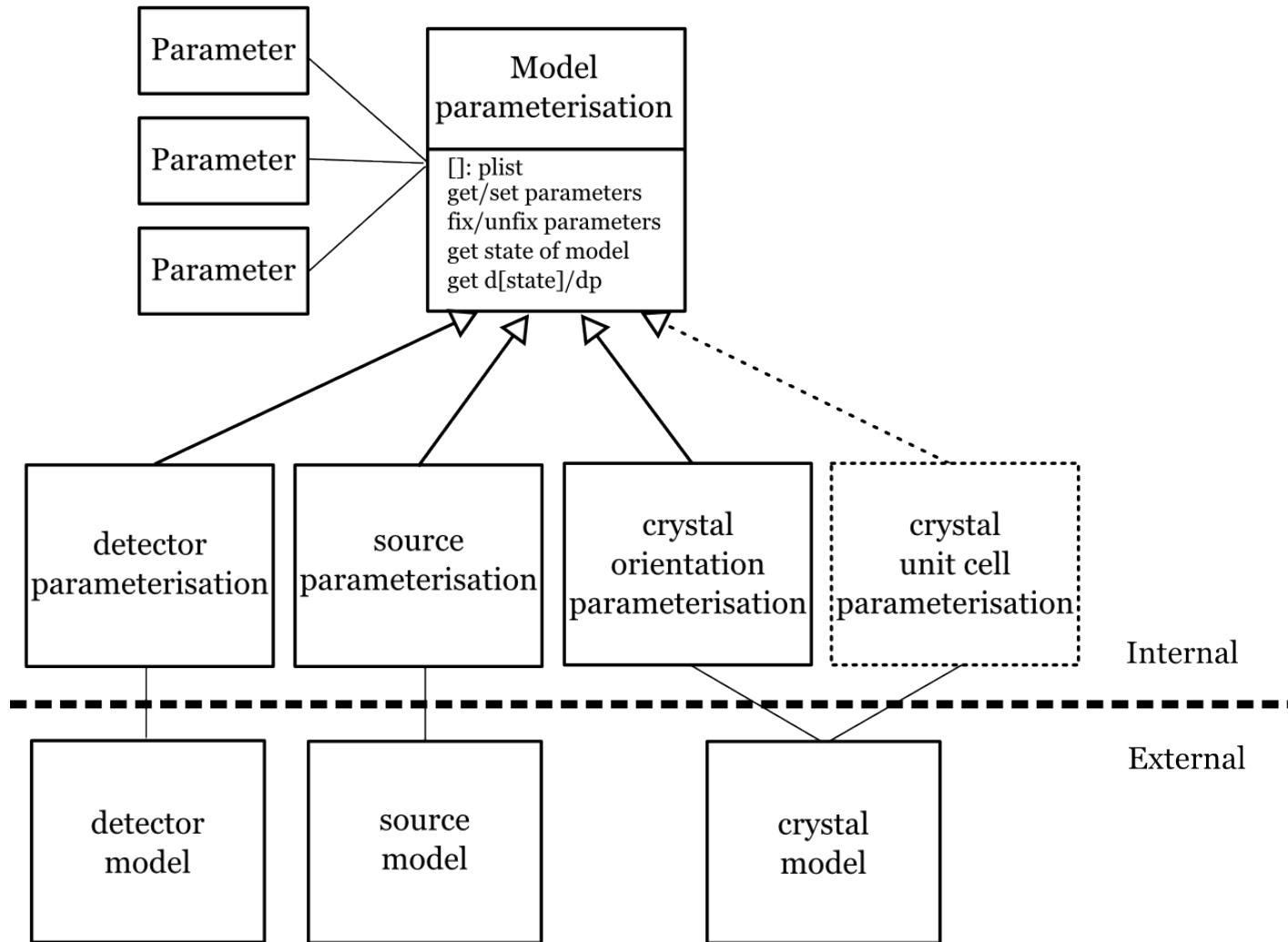
# Parameterisation

Each model parameterisation provides  $\partial[\text{state}]/\partial p$



- Separate 'parameterisation of prediction equation' object takes  $\partial[\text{state}]/\partial p$  for each model and converts to derivatives of  $X$ ,  $Y$ ,  $\phi$  for each reflection
- Individual model parameterisations are encapsulated

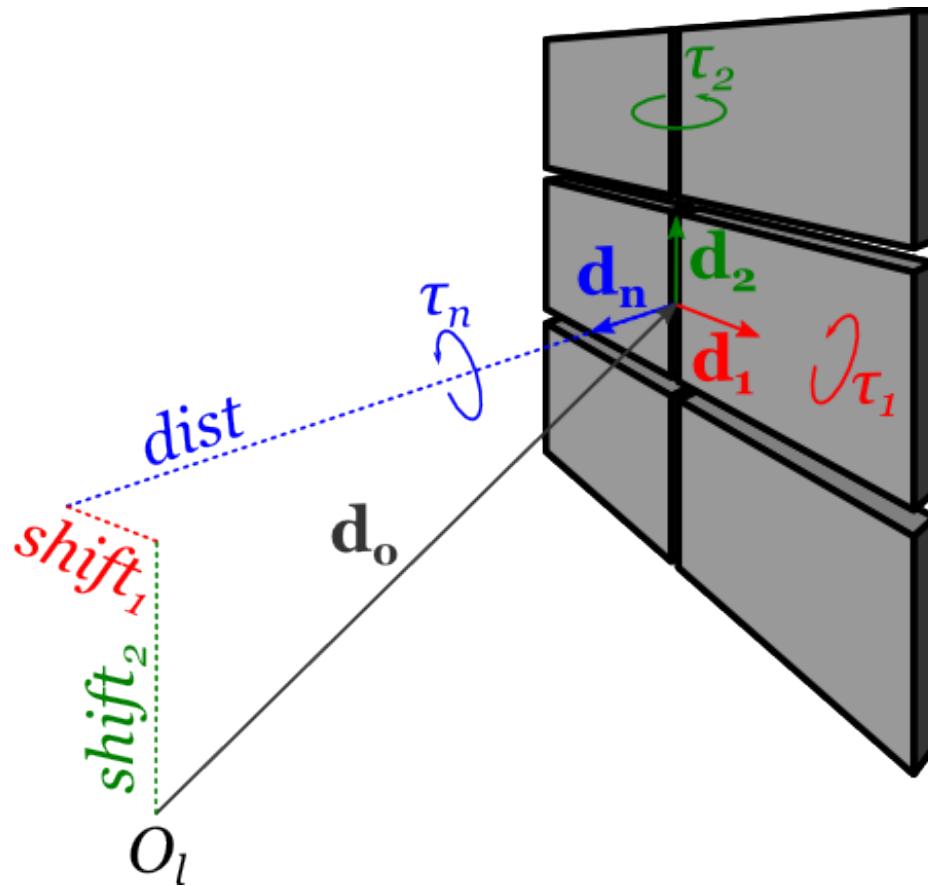
## Parameterisation of experimental models



The abstract interface specifies that:

- Model parameterisations are initialised with an initial state of the model
- New states are composed by the action of functions of the parameters on the initial state
- A state and its derivatives are either a vector or a matrix
- The parameters are either distances or angles with associated unit directions

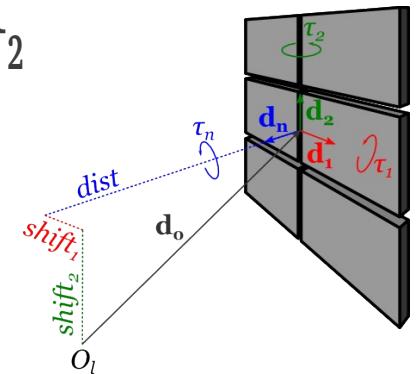
- A concrete example: detector parameterisation



### Parameters

dist - detector distance along  $\mathbf{d}_n$   
shift<sub>1</sub> - translation in direction  $\mathbf{d}_1$   
shift<sub>2</sub> - translation in direction  $\mathbf{d}_2$   
 $\tau_n$  - rotation around  $\mathbf{d}_n$   $\cong$  "roll"  
 $\tau_1$  - rotation around  $\mathbf{d}_1$   $\cong$  "tilt"  
 $\tau_2$  - rotation around  $\mathbf{d}_2$   $\cong$  "twist"

- Initial sensor matrix provides  $\mathbf{d}_0$ ,  $\mathbf{d}_1$ ,  $\mathbf{d}_2$ ,  $\mathbf{d}_n$
- Translation parameters are immediately *dist* along  $\mathbf{d}_n$  and *shift*,  $shift_1$ ,  $shift_2$  along  $\mathbf{d}_1$ ,  $\mathbf{d}_2$
- Initial rotation angles all 0.0, around axes  $\mathbf{d}_1$ ,  $\mathbf{d}_2$ ,  $\mathbf{d}_n$



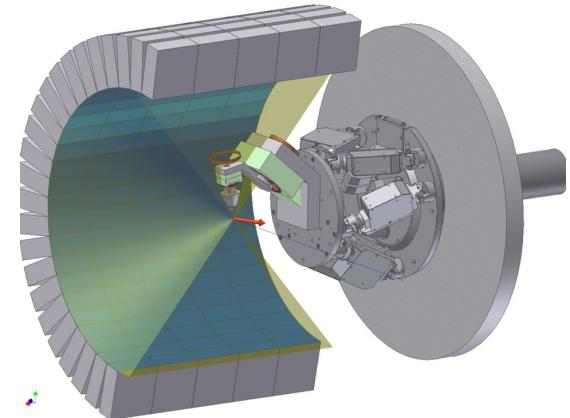
# Parameterisation

- Accommodates refinement of multi-tile detectors as one rigid unit
- Each sensor panel  $k$  has its own matrix  $\mathbf{d}^k = (\mathbf{d}_x^k | \mathbf{d}_y^k | \mathbf{d}_0^k)$
- These vectors are linear combinations of  $\mathbf{d}_0$  and the local coordinate system  $\mathbf{d}_1, \mathbf{d}_2, \mathbf{d}_n$  that moves with the detector:

$$\mathbf{d}_x^k = \alpha_1^k \mathbf{d}_1 + \alpha_2^k \mathbf{d}_2 + \alpha_3^k \mathbf{d}_n$$

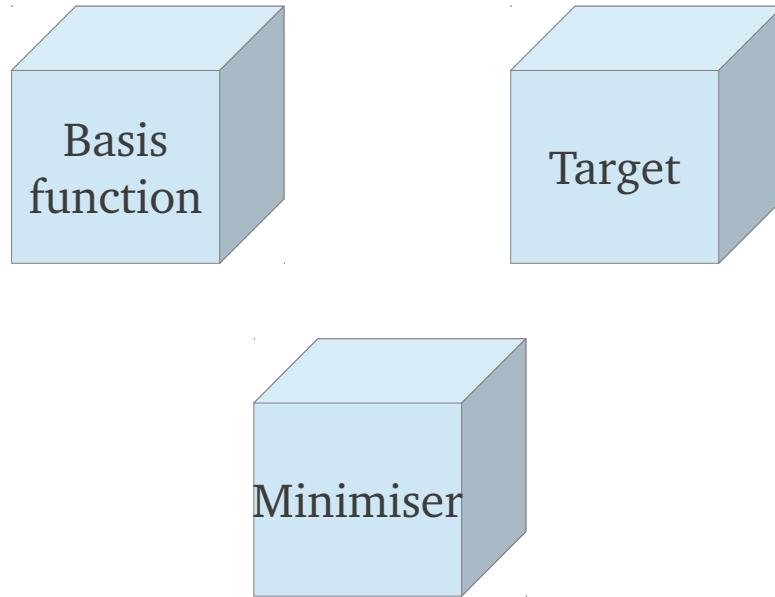
$$\mathbf{d}_y^k = \beta_1^k \mathbf{d}_1 + \beta_2^k \mathbf{d}_2 + \beta_3^k \mathbf{d}_n$$

$$\mathbf{d}_0^k = \mathbf{d}_0 + \gamma_1^k \mathbf{d}_1 + \gamma_2^k \mathbf{d}_2 + \gamma_3^k \mathbf{d}_n$$



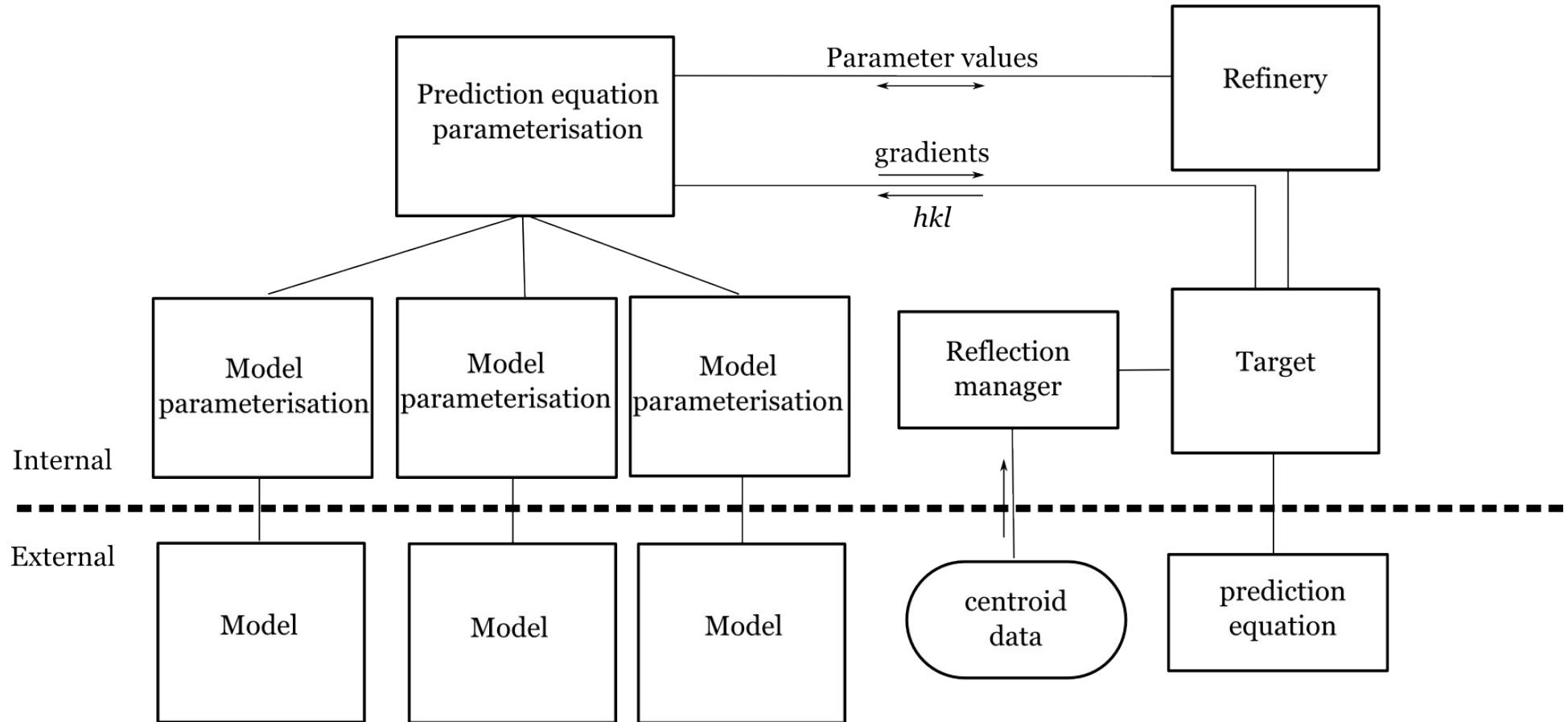
- Thus the derivatives  $\partial \mathbf{d}^k / \partial p$  for each sensor are easily calculated by linear combinations of  $\partial \mathbf{d}_0 / \partial p, \partial \mathbf{d}_1 / \partial p, \partial \mathbf{d}_2 / \partial p$  and  $\partial \mathbf{d}_n / \partial p$

- Further encapsulation within refinement module



- Make these independent (where possible)

## Overview



# Target function

- Least squares target, normalised by the number of observed-predicted pairs currently in play
- The reflections included in refinement are controlled by a reflection manager object

$$L = \frac{1}{2n_h} \sum_h w_{x,h} (X_c - X_o)^2 + w_{y,h} (Y_c - Y_o)^2 + w_{\phi,h} (\phi_c - \phi_o)^2$$

$$\frac{dL}{dp} = \frac{1}{n_h} \sum_h w_{x,h} (X_c - X_o) \frac{dX_c}{dp} + w_{y,h} (Y_c - Y_o) \frac{dY_c}{dp} + w_{\phi,h} (\phi_c - \phi_o) \frac{d\phi_c}{dp}$$

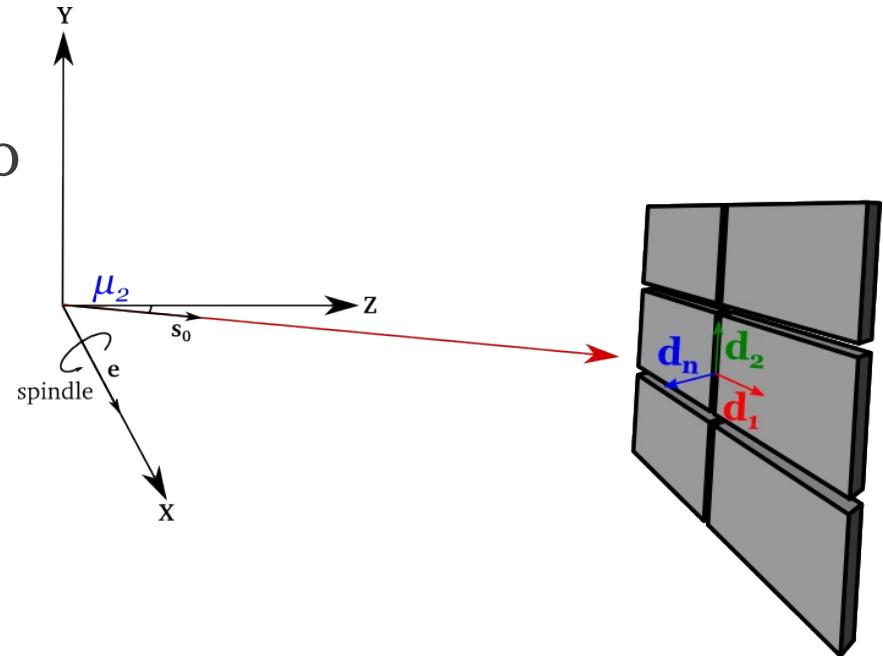
- First order approximation to the curvatures

$$\frac{d^2 L}{dp^2} \approx \frac{1}{n_h} \sum_h w_{x,h} \left( \frac{dX_c}{dp} \right)^2 + w_{y,h} \left( \frac{dY_c}{dp} \right)^2 + w_{\phi,h} \left( \frac{d\phi_c}{dp} \right)^2$$

- So far, only tried scitbx.lbfgs treating it as a “black box”
- Two modes: 1. simple. 2. with curvatures
- Strongly sensitive to parameter scales!

## Protocol

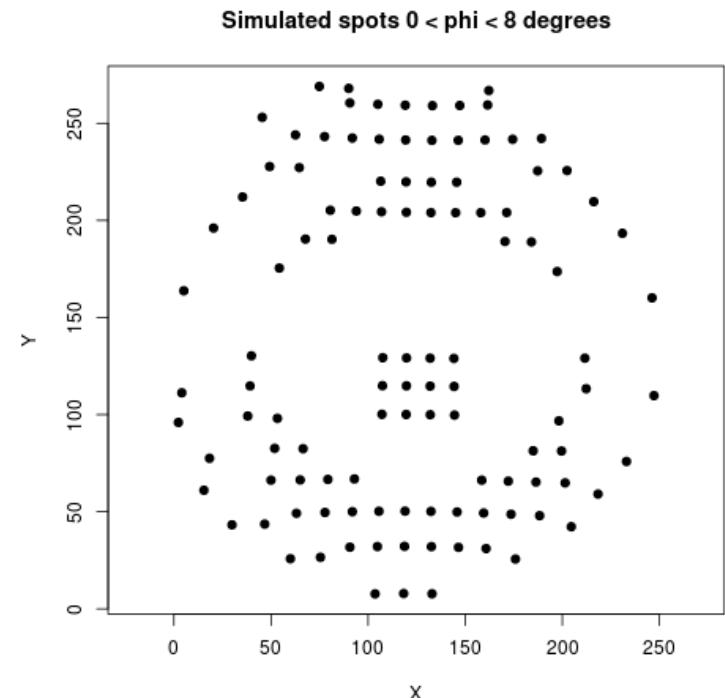
- Generate random geometry, close to an ideal (imgCIF frame)
- Generate a small, random, P1 cell (19.6, 16.7, 15.8, 90.2, 90.2, 89.5)
- Select wavelength (here 1.18 Å)
- Pilatus 2M sized flat panel, ~200 mm from the crystal
- Parameters upset by small amounts from initial values
  - Detector translations shifted 1 mm
  - All rotations (3 detector, 1 beam, 3 crystal) shifted 2 mrad
  - Unit cell fixed and perfect



# Initial test case

- For 180° sweep, 2 Å resolution limit,  $\sim 2.5\text{K}$  'observations' generated by reflection prediction code
- Parameters reset to initial values
- Four runs of refinement attempted

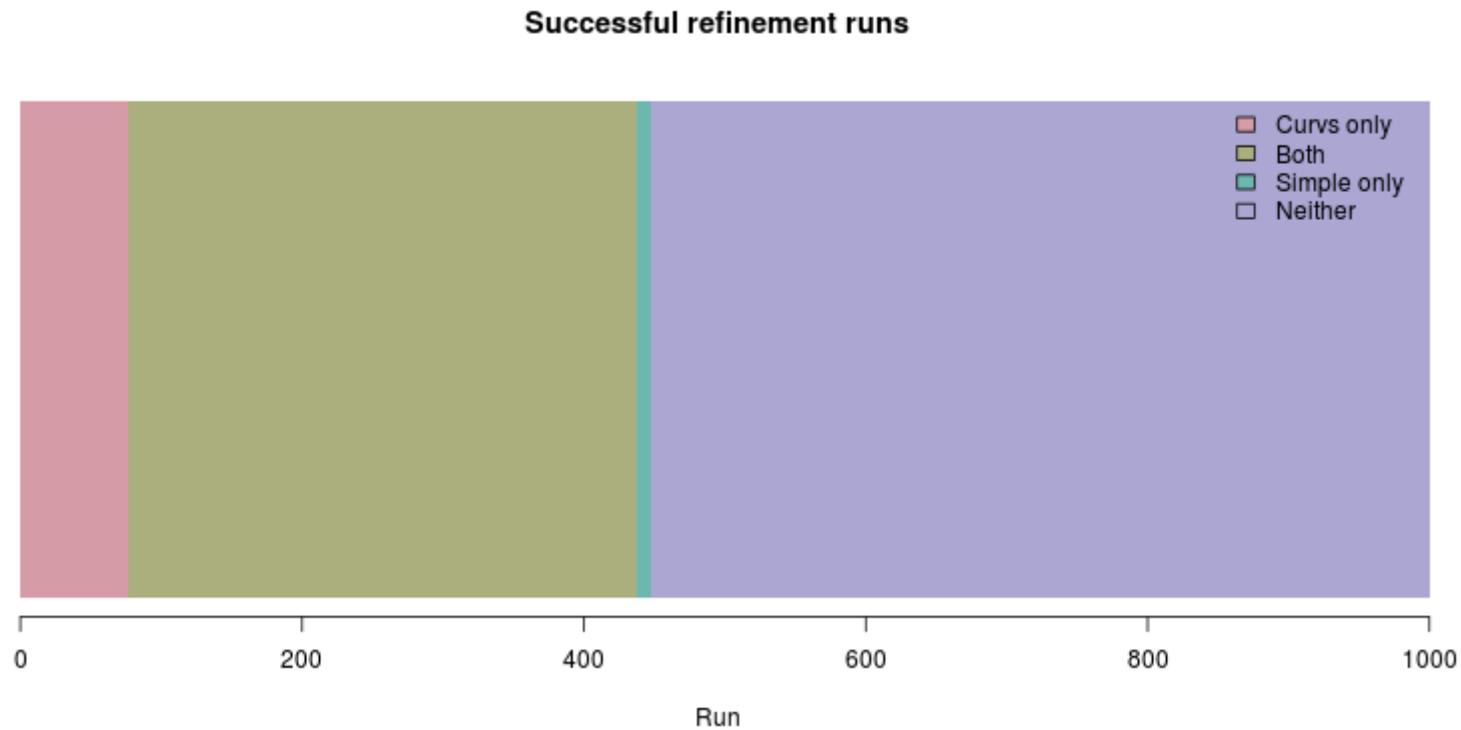
Number of steps to convergence		
	<i>simple</i>	<i>curvatures</i>
<i>mm, rad scale</i>	148	"Line search failed" after 3
<i>mm, mrad scale</i>	15	11



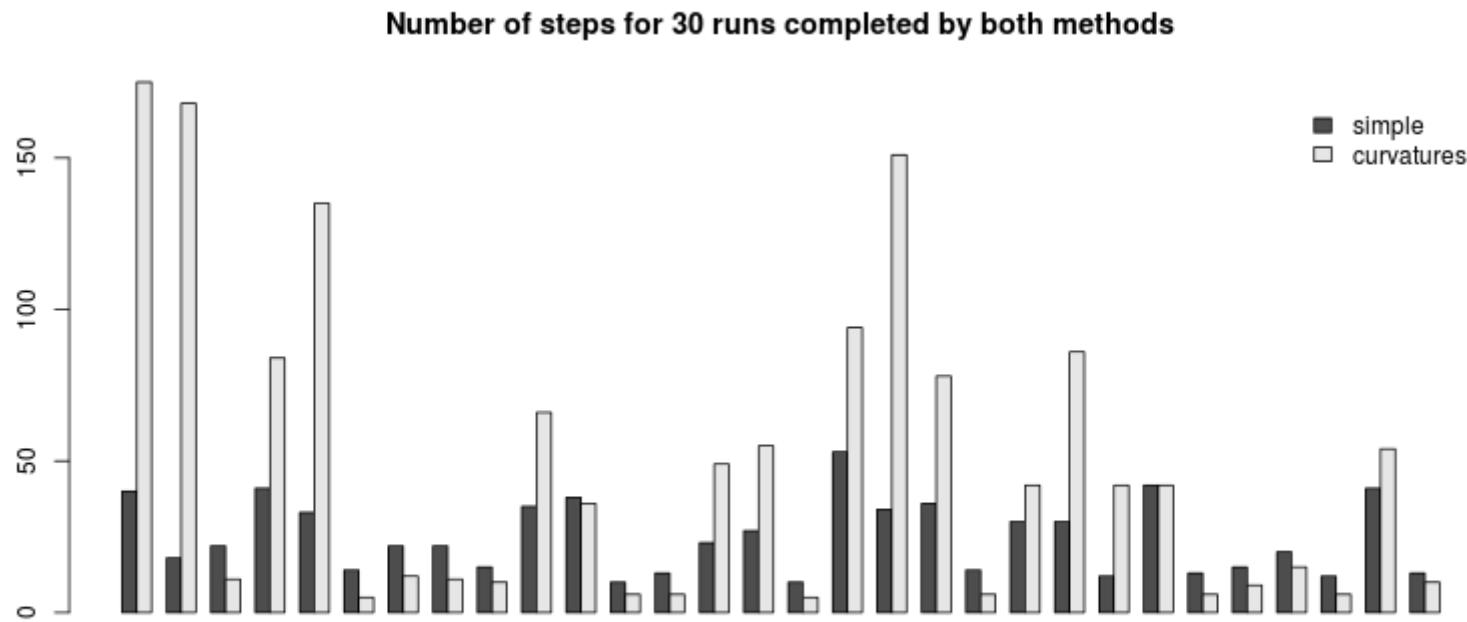
- Initial tests raised a number of problems
  - Original parameter scaling very poor. Problems with line search
  - Tried approximate curvatures to improve performance
  - Likely highly correlated parameters ( $\rightarrow$  large off-diagonal elts of normal matrix, requires investigation)
  - The problem is inherently bounded. Better to have a minimiser that accepts bounds
  - The number of parameters is small, and the target is least squares. More appropriate algorithms than L-BFGS exist ( $\rightarrow$  Gauss-Newton, Levenberg-Marquardt)
  - No time-dependent parameterisation attempted yet; left until other problems are fixed first

- Investigate convergence properties over 1000 tests
- Similar protocol as before:
  - Near imgCIF ideal geometry, small P1 cell (11.3, 16.5, 18.4, 89.6, 90.3, 90.1), beam 0.89 Å. Pilatus 2M sized flat panel ~200 mm away
  - Parameter shifts by normal deviates;  $\sigma=2.0$  mm each translation,  $\sigma=4$  mrad each rotation angle
  - 180° sweep, 2 Å resolution limit, ~1.6K 'observations' generated by reflection prediction code
  - Parameters reset to initial values
  - Refinement attempted to acceptable rmsd

- Results for 1000 attempted runs

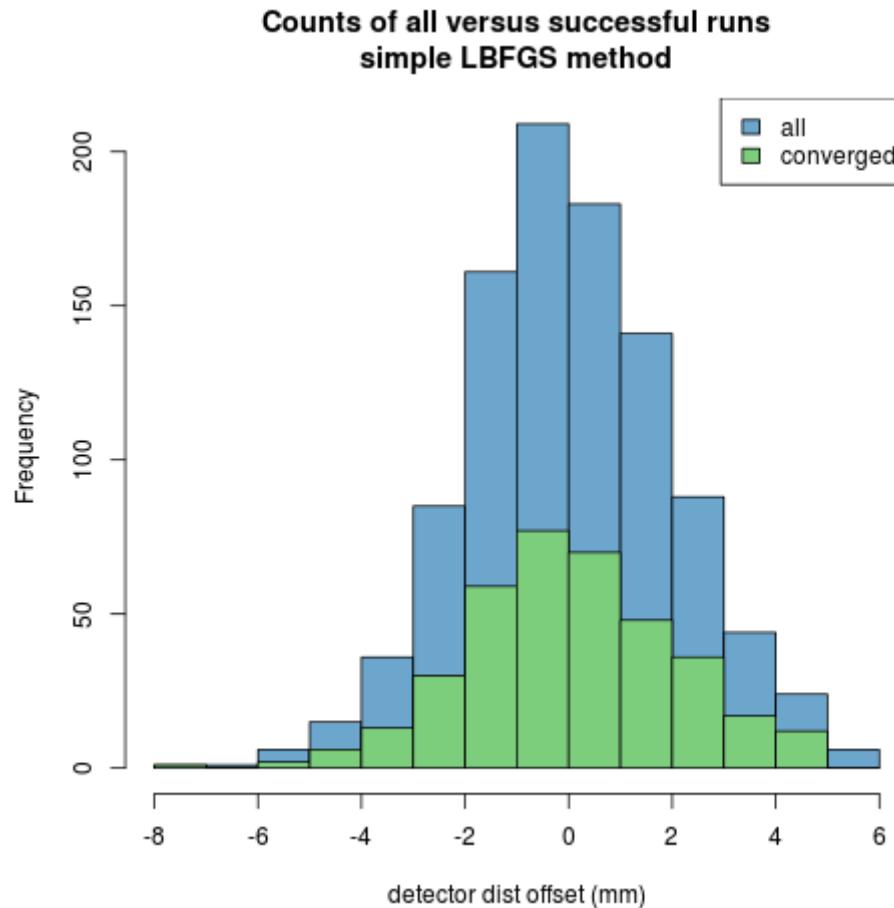


- Using curvatures is surprisingly (?) less efficient for the more difficult runs



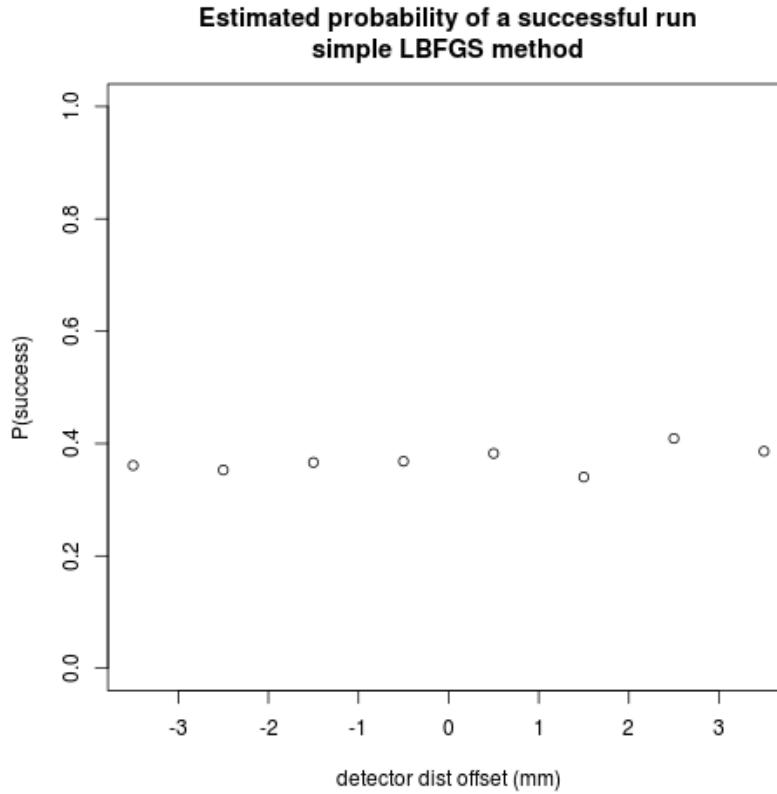
# Convergence radius

- Compare all runs with those successful, one parameter at a time
- Only looking at results for the “simple L-BFGS” method



# Convergence radius

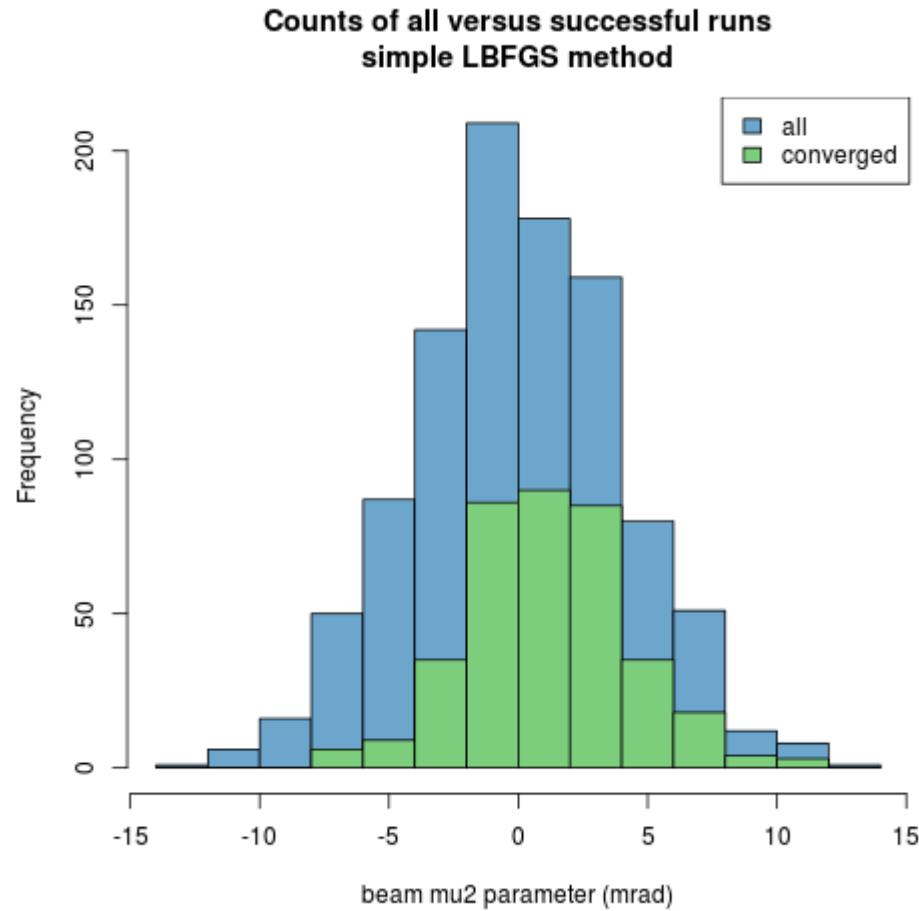
- Compare all runs with those successful, one parameter at a time



- Did not appear to find a failure point for translation parameters

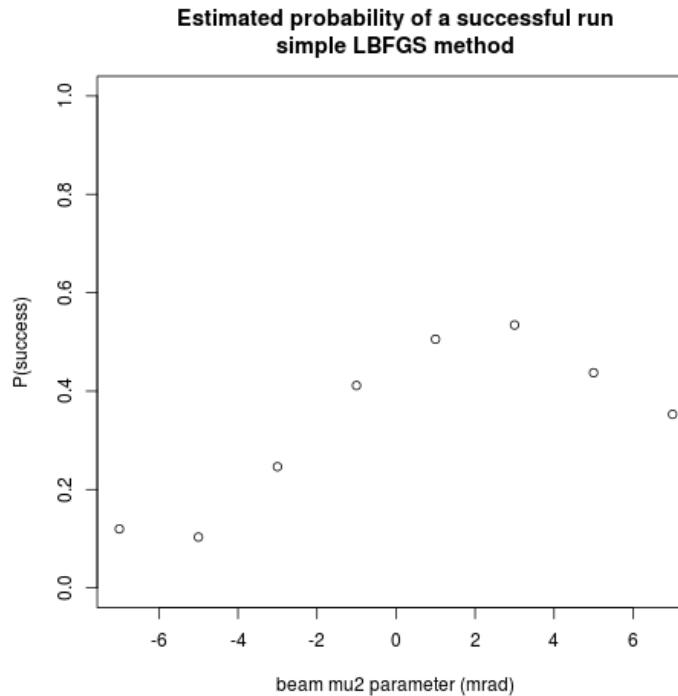
# Convergence radius

- Beam inclination showed most sensitivity



# Convergence radius

- Beam inclination showed most sensitivity



- Not sure why there is a shift of peak success away from the origin
- This investigation is difficult! Is there an established method?

- Include crystal unit cell parameterisation
- Explore different (LS-specific) minimisation engines
- Test against real data
- Time-dependent parameterisation for the crystal; mix global/time-varying refinement
- Multi-sensor detector
- Faster implementation: C++ and parallelism

## Interim conclusions

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- The basic goal of approaching diffraction geometry refinement in a general and modular way is feasible
- Despite some problems with use of the minimiser, the synthetic test cases show promise
- Global refinement with time-varying crystal parameters seems to be an achievable goal
- The “learning by doing” approach breaks the problem down into smaller achievable goals. We are learning much about the nature of the components and interfaces as we go!

# Acknowledgements

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- The CCP4 core team for my time
- Invaluable contributions from Andrew Leslie at the start
- Continuing assistance from a panel of experts: Andrew Leslie, Garib Murshudov, Phil Evans (MRC/LMB), Gleb Bourenkov (EMBL)
- DIALS-East and DIALS-West for pooling resources
- The cctbx community for openness and responsiveness (which is even better than documentation)

# dials

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11 February 2013

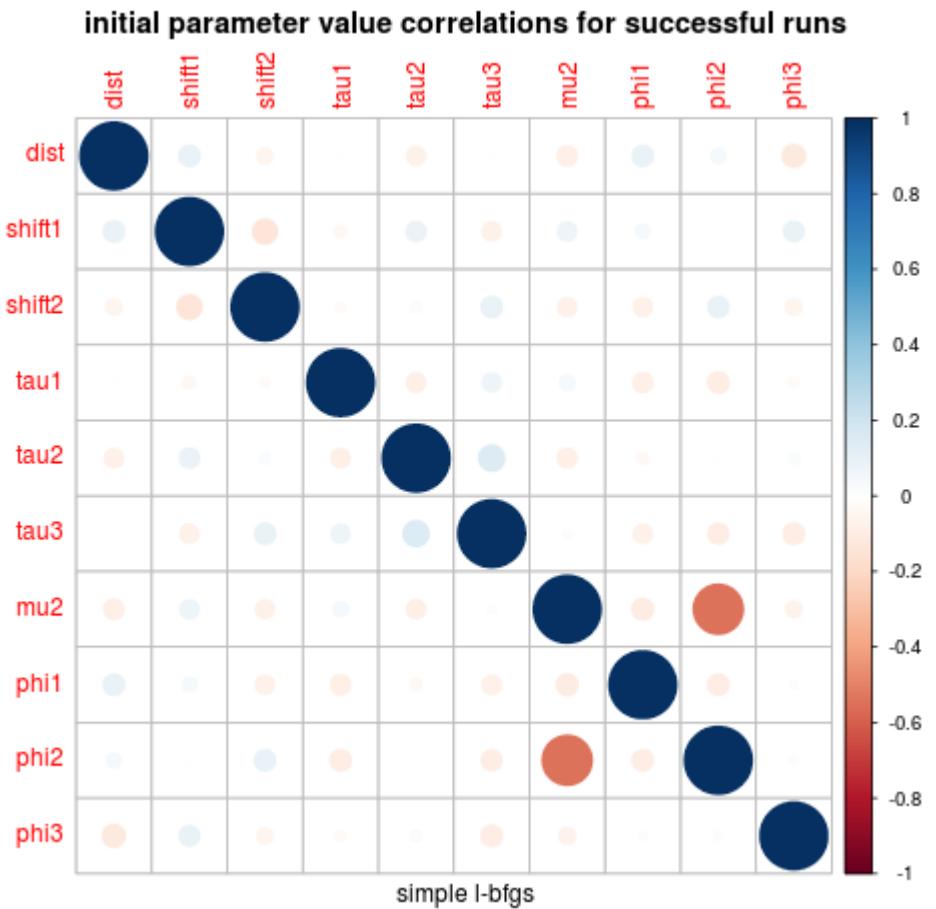


## Extra slides

# Radius of convergence

simple L-BFGS method

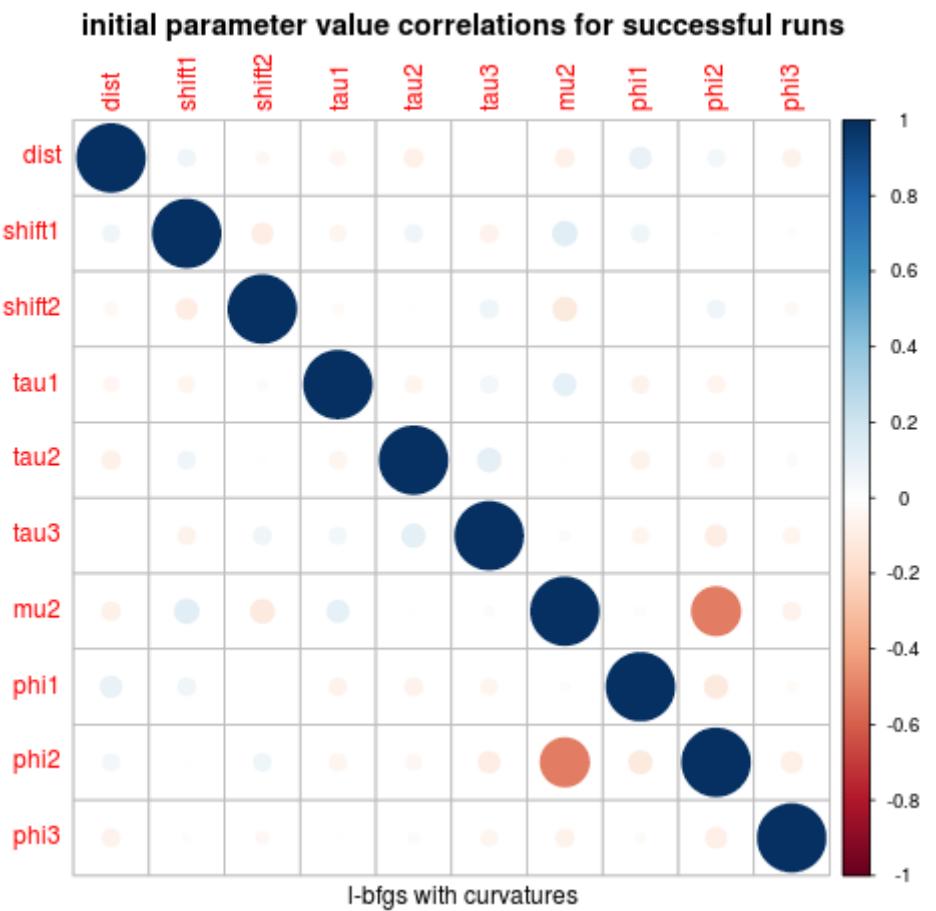
Parameter	$\sigma_{\text{all}}$	$\sigma_{\text{conv}}$	$\sigma_{\text{conv}}/\sigma_{\text{all}}$
dist	0.2	0.2	0.99
shift1	0.2	0.21	1.05
shift2	0.2	0.19	0.97
tau1	3.9	3.59	0.92
tau2	3.8	3.58	0.94
tau3	3.91	3.85	0.99
mu2	4.09	3.25	0.8
phi1	4.01	3.34	0.83
phi2	4.03	3.15	0.78
phi3	4.01	3.74	0.93



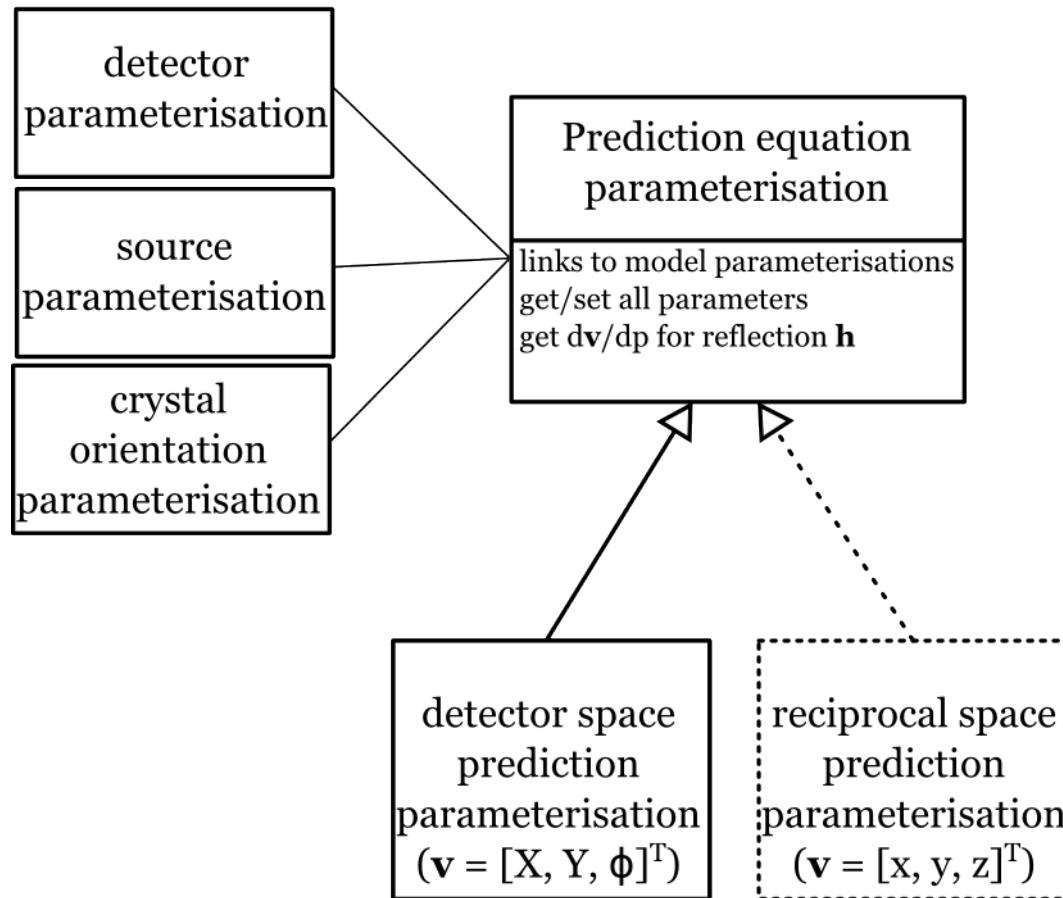
# Radius of convergence

L-BFGS with curvatures

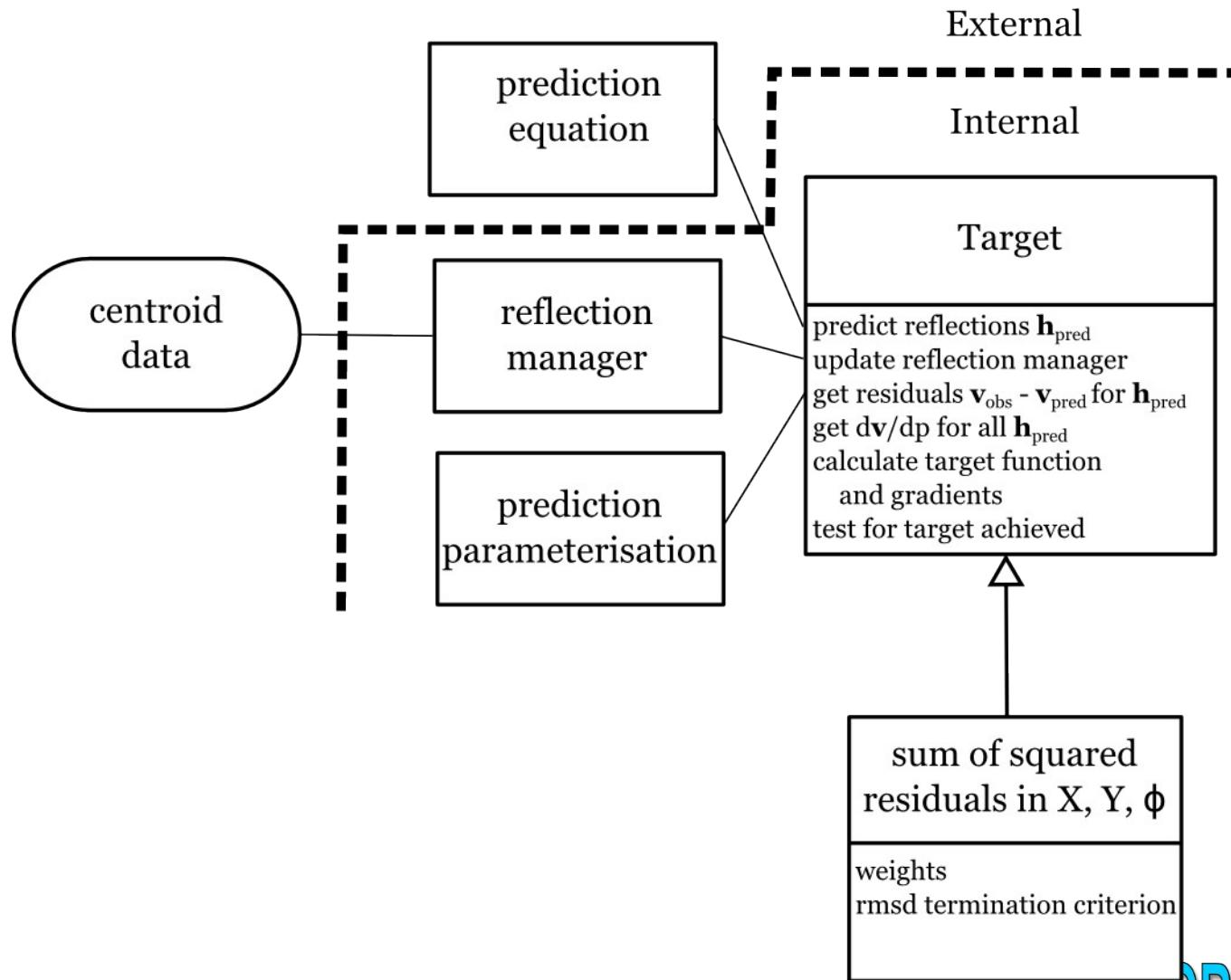
Parameter	$\sigma_{\text{all}}$	$\sigma_{\text{conv}}$	$\sigma_{\text{conv}}/\sigma_{\text{all}}$
dist	0.2	0.2	0.99
shift1	0.2	0.2	1.03
shift2	0.2	0.19	0.97
tau1	3.9	3.84	0.99
tau2	3.8	3.67	0.97
tau3	3.91	3.74	0.96
mu2	4.09	3.44	0.84
phi1	4.01	3.33	0.83
phi2	4.03	3.21	0.8
phi3	4.01	3.96	0.99



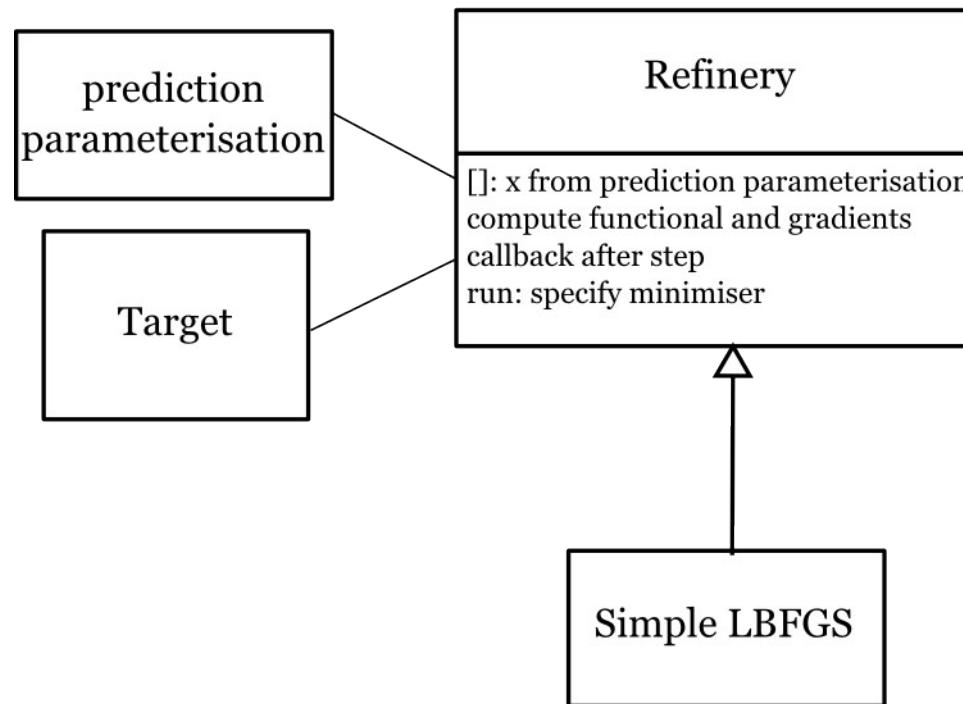
## Parameterisation of the prediction equation



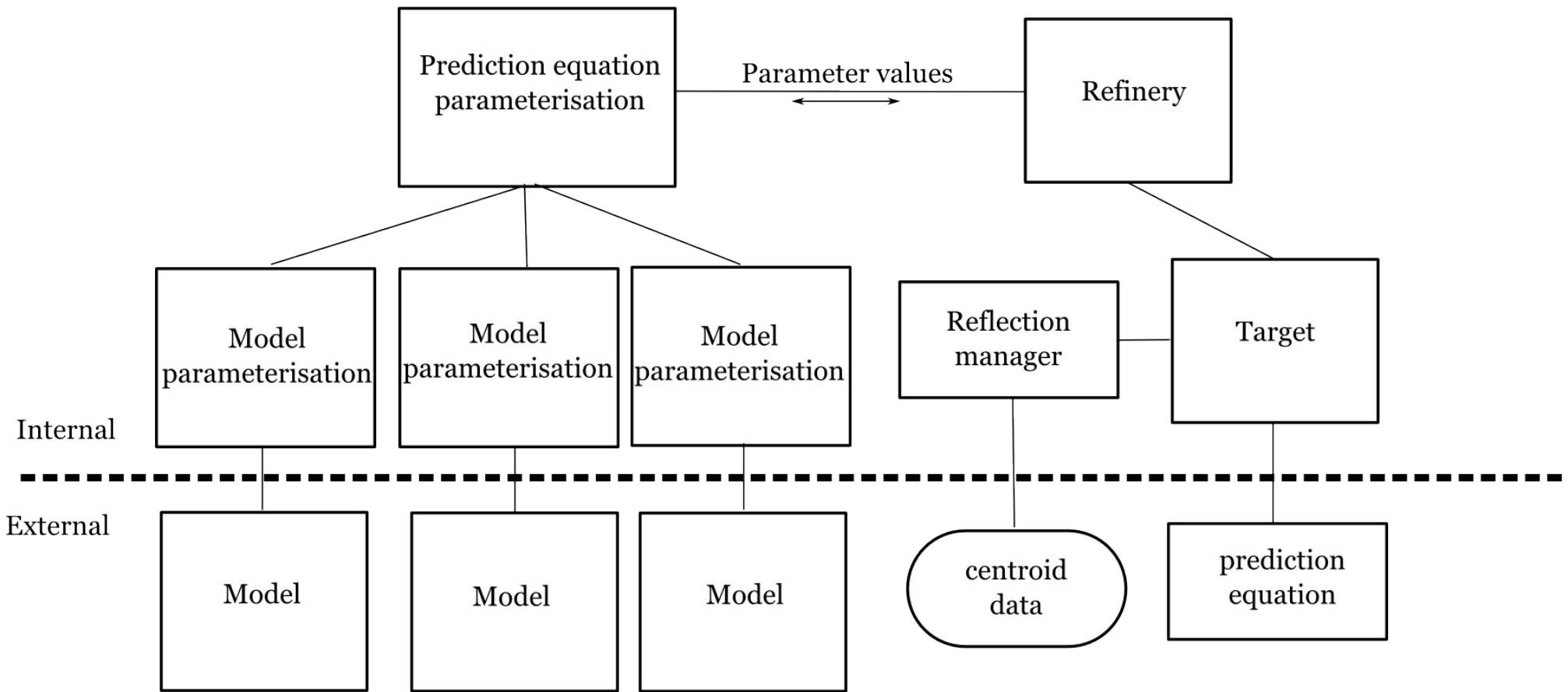
## Data handling and target function



## Refinement engine



## Overview



- “Line search failed” error.

```
lbfgs minimizer step
lbfgs minimizer.run(): f=0.511684, |g|=9.81135, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0467172, |g|=42.5567, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0464431, |g|=42.0279, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.514731, |g|=20.3204, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0463982, |g|=41.9407, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0463332, |g|=41.8142, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.518362, |g|=30.9968, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0463223, |g|=41.793, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0463064, |g|=41.762, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.520692, |g|=36.469, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0463037, |g|=41.7568, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0462998, |g|=41.7491, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.521985, |g|=39.2191, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0462991, |g|=41.7478, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0462981, |g|=41.7459, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0456016, |g|=40.3669, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0451537, |g|=39.458, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0450036, |g|=39.1493, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.522011, |g|=39.2732, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer.run(): f=0.0450036, |g|=39.1493, x_min=-18.0174, x_mean=31.6191, x_max=199.999
lbfgs minimizer exception: lbfgs error: Line search failed: Rounding errors prevent further progress.
There may not be a step which satisfies the sufficient decrease and curvature conditions. Tolerances
may be too small.
lbfgs minimizer done.
```

# Problems

- “Line search failed”

